Total Questions: 21



KNN/Regularization (Q8L9)

Most Correct Answers: #13 Least Correct Answers: #15

- 1. The bias of an estimator (e.g. z^) equals...Hint: the OLS coefficients are unbias:)
- 5/9 A E(z^) z
- 1/9 (B) E(z^2) [E(z)]^2
- 2/9 (C) [E(z^2) E(z)]^2
- 0/9 (D) E(z^2)
- 0/9 (E) I do not know
- 2. The main idea of regularization is
- 4/9 A To introduce a small amount of bias in order to have less variance.
- 2/9 B To introduce a small amount of variance in order to have less bias.
- 2/9 (C) To introduce a small amount of variance and bias in order to have less bias.
- 0/9 D I do not know
- 3. How the tune of any parametr can be made
- 4/9 A using Cross validation
- 0/9 (B) It is impossible
- 0/9 (c) I do not now
- 3/9 (D) using larger sample
- 1/9 $\stackrel{\text{E}}{\bigcirc}$ only having population
- 4. The ridge coefficient estimates shrink towards zero
- 3/9 A when lambda increases
- 4/9 (B) when lambda decreases
- 1/9 $\binom{C}{}$ when lambda = 0
- 0/9 D I do not know

| 5. | Which one can shrink the slope all the way to 0? | | | |
|-----|--|--|--|--|
| 4/9 | A Lasso | | | |
| 4/9 | B Ridge | | | |
| 0/9 | C Regression | | | |
| 0/9 | D I do not know | | | |
| 6. | When lambda = 0, we have | | | |
| 2/9 | (A) Ridge | | | |
| 4/9 | B Lasso | | | |
| 0/9 | © EL | | | |
| 2/9 | D Regression | | | |
| 0/9 | E I do not know | | | |
| 7 | When alpha - 0 we have | | | |
| 7. | When alpha = 0, we have | | | |
| 1/9 | A Ridge | | | |
| 7/9 | (B) Lasso | | | |
| 0/9 | C EL | | | |
| 0/9 | (D) Regression | | | |
| 0/9 | E I do not know | | | |
| 8. | Which function can help to perform cross-validation for regularization in R? | | | |
| 5/9 | A cv.glmnet() | | | |
| 0/9 | B cros_val() | | | |
| 3/9 | © glmnet(method = "cv) | | | |
| 1/9 | D I do not know | | | |
| 9. | KNN is | | | |
| 4/9 | A Data-driven | | | |
| | B Model-driven | | | |
| 2/9 | | | | |
| 2/9 | C I do not now | | | |
| | | | | |

| 10. | KN | N is |
|-----|--------------|---|
| 1/9 | A | parametric method |
| 5/9 | В | non-parametric method |
| 2/9 | С | I do not know |
| 11. | The | e dependent variable of the (OLS) regression is |
| 1/9 | A | categorical |
| 0/9 | B | ordinal |
| 5/9 | C | continuous |
| 2/9 | D | count |
| 1/9 | E | I do not know |
| 12. | The | e dependent variable of the classification is |
| 5/9 | A | categorical |
| 2/9 | B | numeric |
| 1/9 | C | I do not know |
| 13. | Но | w to chose K? |
| 0/9 | \bigcirc A | pick own |
| 7/9 | В | using cross-validation |
| 1/9 | \bigcirc | the largest one |
| 1/9 | D | the smallest one |
| 14. | KN | N can be used for regression |
| 4/9 | A | Yes |
| 1/9 | B | No |
| 3/9 | C | I do not know |
| 15. | In 1 | the case of KNN classification we use |
| 4/9 | \bigcirc A | average of outcomes |
| 1/9 | В | majority voting scheme |
| 3/9 | C | I do not know |

| 16. | Which of these errors will increase constantly by increasing k? |
|-----|---|
| 5/9 | A train error |
| 2/9 | B test error |
| 0/9 | C both |
| 1/9 | D I do not know |
| 17 | This formation can be used to newform (ANN in D |
| 17. | This function can be used to perform KNN in R |
| 2/9 | A knn() |
| 1/9 | B k_nn() |
| 1/9 | C knnreg() |
| 1/9 | D knearneib() |
| 3/9 | E I do not know |
| | |
| 18. | With the increase of k, the decision boundary will be |
| 2/9 | A simplified |
| 0/9 | B more complex |
| 4/9 | C I do not know |
| 2/9 | D unchanged |
| | |
| 19. | The best k correspond to |
| 5/9 | A the lowest point of test error |
| 1/9 | (B) the lowest point of train error |
| 0/9 | C the highest point of test error |
| 2/9 | D I do not know |
| 20 | KNINI algorithm is consitive to entliers |
| 20. | KNN algorithm is sensitive to outliers |
| 3/9 | A True |
| 3/9 | B False |
| 2/9 | C I do not know |
| | |

- 21. KNN
- 5/9 A is a supervised learning algorithm.
- 2/9 B is an unsupervised learning algorithm.
- 1/9 C I do not know